Class 5: Data Visualization

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Base R graphics vs ggplot2

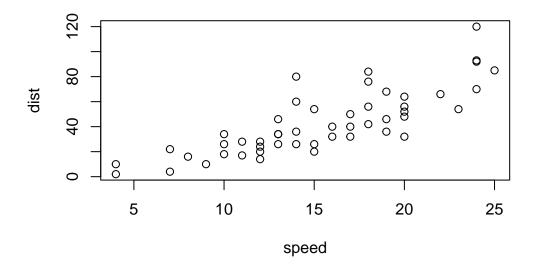
There are many graphics systems available in R, including so-called "base" R graphics and the very popular **ggplot2** package.

To compare these let's play with the inbuilt cars dataset.

```
head(cars)
```


To use "base" R I can simply call the plot() functions:

```
plot(cars)
```



To use ggplot2 package I first need to install it with the function install.packages("ggplot2").

I will run this in my R console (i.e. the R brain) as I do not want to re-install it every time I render my report.

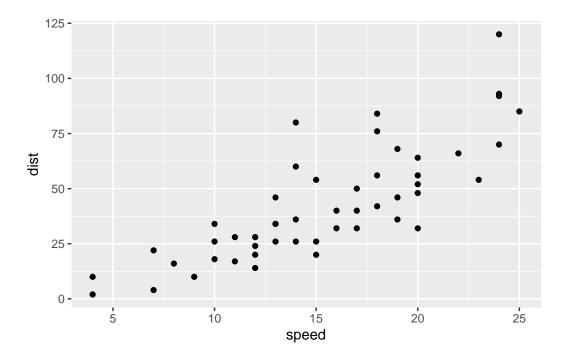
The main function in this package is called ggplot(). Can I just call it?

```
library(ggplot2)
ggplot()
```

To make a figure with ggplot I need always at least 3 layers:

- data (i.e. what I want to plot)
- \mathbf{aes} (the aesthetic mapping of the data to the plot I want)
- the **geoms** (i.e. How I want to plot the data)

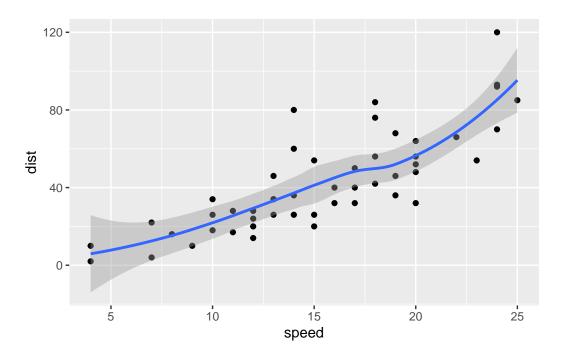
```
ggplot(data=cars) +
  aes(x=speed, y= dist) +
  geom_point()
```



If I want to add more things I can just keep adding layers, e.g.

```
ggplot(data=cars) +
  aes(x=speed, y= dist) +
  geom_point() +
  geom_smooth()
```

 $[\]ensuremath{\mbox{`geom_smooth()`}}\ \mbox{using method} = \ensuremath{\mbox{'loess'}}\ \mbox{and formula} = \ensuremath{\mbox{'y}}\ \sim \ensuremath{\mbox{x'}}\ \mbox{'}$



GGplot is much more verbose than base R plots for standard plots but it has a consistent layer system that I can use to make just about any plot.

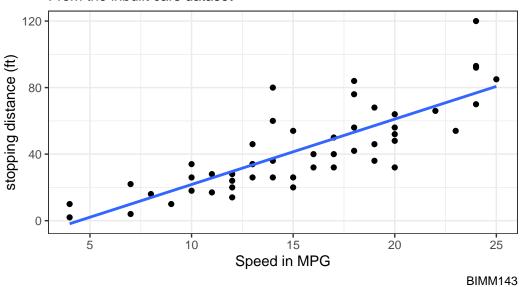
Let's make a plot with a straight line fit - i.e. a linear model and no standard error shown.

```
ggplot(data=cars) +
  aes(x=speed, y= dist) +
  geom_point() +
  geom_smooth(se=FALSE, method="lm") +
  labs(title = "Stopping distance for old cars", subtitle = "From the inbuilt cars dataset
  theme_bw()
```

[`]geom_smooth()` using formula = 'y ~ x'

Stopping distance for old cars

From the inbuilt cars dataset



A more complicated plot

Let's plot some gene expression data The code below reads the results of a differential expression analysis where a new anti-viral drug is being tested.

```
url <- "https://bioboot.github.io/bimm143_S20/class-material/up_down_expression.txt"
genes <- read.delim(url)
head(genes)</pre>
```

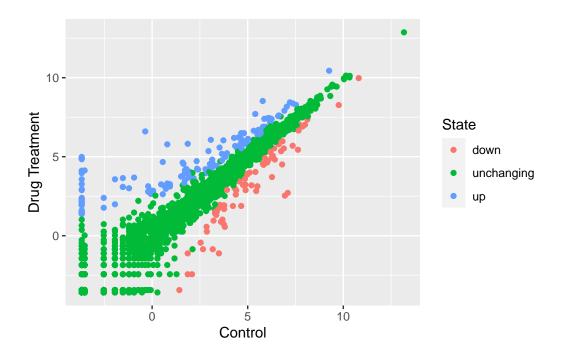
```
Gene Condition1 Condition2 State
A4GNT -3.6808610 -3.4401355 unchanging
AAAS 4.5479580 4.3864126 unchanging
AASDH 3.7190695 3.4787276 unchanging
AATF 5.0784720 5.0151916 unchanging
AATK 0.4711421 0.5598642 unchanging
AB015752.4 -3.6808610 -3.5921390 unchanging
```

Q. How many genes are in this dataset?

```
nrow(genes)
```

```
[1] 5196
    Q. How many columns are in this dataset?
  ncol(genes)
[1] 4
    Q. How can we summarize the last column - the "state" column?
  table(genes$State)
      down unchanging
                               up
        72
                 4997
                              127
  table(genes$State)
      down unchanging
                               up
        72
                 4997
                              127
  round(table(genes$State)/nrow(genes)*100, 2)
      down unchanging
                               up
      1.39
                96.17
                             2.44
  ggplot(genes) +
    aes(x=Condition1, y=Condition2, col=State) +
    geom_point() +
```

labs(x="Control", y="Drug Treatment")

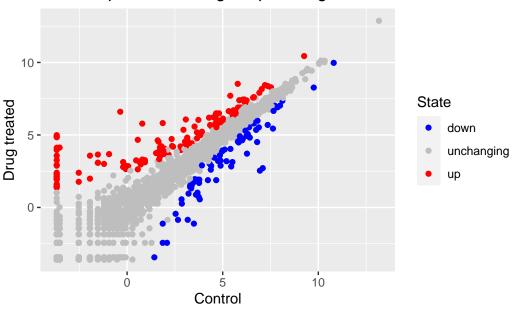


```
p <- ggplot(genes) +
  aes(x=Condition1, y=Condition2, col=State) +
  geom_point()</pre>
```

I can now just call ${\tt p}$ when I want to plot or add to it.

```
p + labs(title="Gene Expression changes upon drug treartment", x="Control", y="Drug treate
scale_color_manual( values = c ("blue", "gray", "red"))
```

Gene Expression changes upon drug treartment

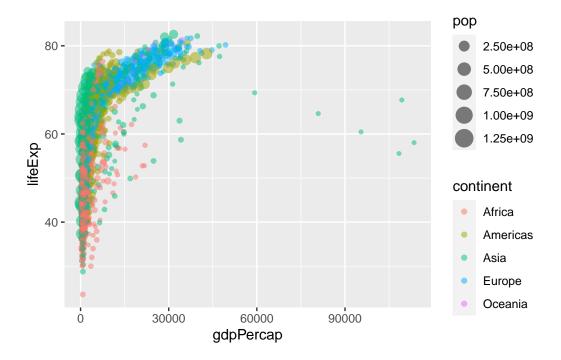


Going Further

Here I read a slightly larger dataset

```
# File location online
  url <- "https://raw.githubusercontent.com/jennybc/gapminder/master/inst/extdata/gapminder.
  gapminder <- read.delim(url)</pre>
  head(gapminder)
      country continent year lifeExp
                                         pop gdpPercap
1 Afghanistan
                   Asia 1952 28.801 8425333 779.4453
2 Afghanistan
                  Asia 1957 30.332 9240934 820.8530
3 Afghanistan
                  Asia 1962 31.997 10267083 853.1007
4 Afghanistan
                  Asia 1967 34.020 11537966 836.1971
5 Afghanistan
                  Asia 1972 36.088 13079460 739.9811
6 Afghanistan
                  Asia 1977 38.438 14880372 786.1134
```

```
ggplot(gapminder) +
  aes(x=gdpPercap, y=lifeExp, color=continent, size=pop) +
  geom_point(alpha=0.5)
```



A very useful layer to add sometimes is for "faceting"

```
ggplot(gapminder) +
  aes(x=gdpPercap, y=lifeExp, color=continent, size=pop) +
  geom_point(alpha=0.3) +
  facet_wrap(~continent)
```

